Unpacking anxiety in climate adaptation: a digital media analysis of public discourse

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### Introduction

**Overview:** In the era of accelerating climate change, digital platforms like YouTube are not just sources of information, they are spaces where emotions, narratives, and ideologies collide. As climate adaptation becomes more urgent, understanding how online discourse shapes public sentiment is crucial.

**Context:** Social media platforms significantly shape public perceptions and attitudes toward climate change. They function as arenas of both collective solidarity and polarized debate.

**Research Question:** Does social media discourse around climate change foster societal solidarity or deepen polarization?

**Aim of the Study:** This research examines the emotional dynamics and stances on misinformation in YouTube comment sections. It seeks to uncover how public discourse around climate adaptation may either build collective engagement or trigger division, particularly through expressions of anxiety.

#### Structure

- Part 1: General overview of emotional patterns and polarization in climate discourse
- Part 2: A focused analysis of anxiety-related comments and their social meaning



### Methodology

- 74 English language videos most relevant to the search query of "climate change" with comments extracted from YouTube on April 16, 2025 using YouTube API.
  - Views : 1 100 to 13 714 014 (mean: 1 460 616.54, std: 2 695 756.00);
  - Published from 2014-04-22 to 2025-03-25. Most videos published: 2024-10:7 videos; 2025-03: 7 videos.
- 334 708 comments extracted for these videos.
  - Likes: 0 to 69 416 (mean: 7.26, std: 255.68);
  - Responses: 0 to 756 (mean: 0.68, std: 7.40);
  - Published from 2014-04-22 to 2025-03-25. Top 5 months by comment count: 2021-09: 40 053 comments; 2022-04: 26 837 comments; 2020-06: 25 734 comments; 2024-01: 22 800 comments; 2023-09: 18 569 comments.
- OpenAI API model gpt-4o-mini used to process the comments.
- Potential limitations: representativeness of selected videos, platform-specific trends and algorithmic filtering, limitations of OpenAI API model.



## Part 1: General overview of emotional patterns and polarization in climate discourse









**Misinformation Stance** 



### Sentiment score and emotions



#### Most frequent emotions in polarizing and solidarity comments (in %)

	frustration	skepticism	anger	concern	sarcasm	cynicism	fear	dismissive	discontent	suspicion
polarizing	50.9	15.1	11.1	2.7	3.4	1.9	1.4	1.3	1	0.9

	hope	gratitude	positive	frustration	joy	concern	optimism	agreement	determination
solidarity	18.9	9.5	7.6	6.6	6.2	5.3	5	3.3	3.1

# Misinformation stance

Examples of comments (altered for the purpose of anonymity):

- Supporting: Remember climate change is created by CERN.
- Opposing: You should listen to real scientists.





### Most frequent misinformation narratives

Narrative category	Example claim	%			
Attribution & courselity microprocontation	Climate shifts are entirely natural (sun-cycles, volcanoes, Milanković cycles).	22 22			
Attribution & causality misrepresentation	CO <sub>2</sub> is plant food; more of it is good and can't warm the planet anyway.	27.03			
Climate change denial	There is no climate crisis – global warming is a hoax.	25.77			
Conspiracy theories	Climate is being engineered by chemtrails / HAARP / weather weapons.	24.04			
Impact denial / minimization	The claimed impacts (sea-level rise, reef death, extreme weather) aren't happening.	12.78			
	Scientists and agencies fake the data.				
Delegitimization of science & institutions	Climate change is a political scam to tax and control the public.				
Others		1.88			



Distribution of Solidarity vs. Polarizing Comments



# Solidarity vs. polarizing

Examples of comments (altered for the purpose of anonymity):

- Solidarity: Humanity should protect the ecosystem now!
- Polarizing: Climate change is a fake problem created by the UN. It has always changed.

### Solidarity claims

### Expressions of gratitude and encouragement

- Thank you so much
- Keep up the good/great work
- Thank you for making this video
- I really needed this / I needed to hear this
- Thank you for giving me hope

### Calls for collective action

- We all need to fix it
- We can do this / We can fix it
- Need to work together
- Do something about climate change
- We need more people to act

### Affirmations of climate science

- Climate change is real
- Scientific consensus on climate change
- Human activity is driving climate change

### Solidarity and unity

- We're all in this together
- Every nation on Earth must act
- Part of the solution
- The world a better place

### Hope and optimism

- Hope for the future
- Make a difference
- It's not too late
- Breath of fresh air



## Polarizing claims

### **Climate denial and hoax narratives**

- Climate change is a hoax/myth/scam
- There is no climate crisis/change
- Climate change isn't real
- Man-made climate change is fake

### Natural climate change arguments

- Climate has always been changing
- Climate change is natural
- Coming out of an ice age
- CO<sub>2</sub> is plant food / natural

### **Rejection of human responsibility**

- Has nothing to do with humans
- Man-made climate change isn't real
- Humans aren't responsible
- Nothing to do with CO<sub>2</sub>/fossil fuels

### **Conspiracy and mistrust**

- Sponsored by Bill Gates
- Fossil fuel industry propaganda
- Do your own research

### Skepticism and cynicism

- End of the world hysteria
- It's all about money/control
- Sky is falling panic
- Don't believe in climate change

### Dismissal of climate activism

- They don't care / We don't care
- Not the other way around
- We don't need to do anything
- Nothing we can do about it



### Likes and replies

	Misinfo	rmation stance			Solidarit	y vs. polarizing	
Outcome	F	p	Group means (highest → lowest)	Outcome	F	р	Group means (highest $\rightarrow$ lowest)
Likes Replies	6.54 14.13	0.001 < 0.001	Opposes > Neutral > Supports Supports >	Likes	13.96	< 0.001	Solidarity (9.35) ≫ None (4.12) > Polarizing (1.74)
	2704.00	0.004	Opposes > Neutral	Replies	6.67	0.001	Solidarity > Polarizing > None
Sentiment	2704.90	< 0.001	Neutral (− 0.12) > Supports (− 0.46) ≈ Opposes (− 0.47)	Sentiment	16262.18	< 0.001	Solidarity (+0.26) > None (−0.11) ≫ Polarizing (−0.53)



### Clustering

Cluster	Comments (%)	Sentiment	Likes	Responses	Solidarity_polarizing			Misinformation stance			
					None	Polarizing	Solidarity	Neutral	Opposes	Supports	Unrelated
0	2.50	-0.645	182.797	11.769	95.3	3.2	1.5	0.1	0.6	0.1	99.1
1	5.22	0.71	12.14	0.698	0	0	100	0.1	3.1	0.1	96.7
2	4.36	-0.488	0.622	0.879	7.8	<mark>92.1</mark>	0.1	0	0	<mark>100</mark>	0
3	9.84	0.48	7.257	0.609	97.1	2.9	0	0	0	0	99.9
4	19.54	0.089	1.468	0.251	100	0	0	2.4	0	0.1	97.5
5	7.53	-0.659	2.362	0.488	0	<mark>94.2</mark>	5.8	0	0	0	100
6	16.08	-0.501	1.649	0.637	15.8	<mark>83.7</mark>	0.5	4.8	<mark>95.2</mark>	0	0
7	7.92	0.766	2.748	0.224	99	1	0	0.9	4.7	1.2	93.3
8	19.60	-0.464	1.305	0.129	99.8	0	0.2	0	0	0	100
9	7.41	-0.449	1.498	0.271	0	100	0	0	0	0	100



### Analysis over time





### Authors





## Part 2: A focused analysis of anxiety-related comments and their social meaning



## ✓ Is Anxiety Clearly Distinguishable from Other Emotions?

### **Theoretical clarity**

- Anxiety is **conceptually** distinct from other negative emotions:
- **Fear** is more immediate and specific (e.g., "I'm scared of wildfires this summer").
- **Anger** involves a sense of injustice or blame (e.g., "Politicians are doing nothing!").
- **Frustration** stems from blocked goals (e.g., "Why is this taking so long to fix?").
- Anxiety, in contrast, is more diffuse, futureoriented, and internalized, often expressing worry, dread, or a sense of helplessness (e.g., "I can't stop thinking about how bad things might get").

### • 2. In practice

- In real-world, noisy, user-generated text **anxiety** is often blended with fear and frustration, making it tricky to isolate.
- People may not use the word *anxiety* directly, but instead use **signals** like:
  - $_{\odot}$  "I can't sleep at night thinking about this."
  - $\circ~$  "This gives me a pit in my stomach."
  - $_{\odot}\,$  ``I'm constantly worried about the future."

#### **Misinformation Related Distribution**

#### Solidarity Polarizing Distribution



Misinformation Stance Distribution (Related Comments)





#### Sentiment Score Distribution for Anxiety Comments



0.21% of all comments.

Like count (from 0 to 1881) with mean 8.37 and std. 91.64.

IASA

Response count (from 0 to 57) with mean 0.74 and std. 3.83.



### Anxiety comments vs. other comments

#### **Sentiment score**

- Group comparison: The Mann– Whitney U test is highly significant (p < .001), telling us the distribution of sentiment scores for anxiety comments is different from that of all others.
- Effect size: Cohen's d = -0.639 is a moderate-to-large effect, indicating anxiety comments are on average notably more negative in sentiment than other comments.

#### Like count

 Group comparison: Mann– Whitney p = .065 so we do not have evidence of a reliable difference in like counts between anxiety comments and other comments.

 Effect size: Cohen's d ≈ 0.004, essentially zero, confirms there's no meaningful difference in how much these two groups get liked.

#### Number of responses

- Group comparison: Mann– Whitney p = .041 is just under .05, technically significant.
- Effect size: Cohen's d ≈ 0.009, basically zero.

#### **Misinformation related**

- Contingency: Anxiety comments are misinformation-related in  $64/717 \approx 8.9\%$  of cases; the rest of the corpus is misinfo-related at 72 421/333 991  $\approx$  21.7%.
- Chi-square:  $\chi^2 \approx 67.9$ , p < .001 confirms a real difference in proportions.

• Effect size: Cramér's V = 0.014 is extremely small.

#### **Misinformation stance**

- Chi-square:  $\chi^2 \approx 105.8$ , p < .001 indicates a group difference in stance distributions.
- Effect size: Cramér's V = 0.018 still vanishingly small.

#### Solidarity / Polarizing

- Anxiety comments are coded as "none" \~80%, "polarizing" \~14%, "solidarity" \~6%; for other comments it's \~62%/32%/6%.
  - $\chi^2 \approx 115.2$ , p < .001 shows a significant distributional difference, but Cramér's V = 0.019 again tells us the magnitude is trivial.

### Conclusion

- Sentiment is the lone standout: Anxiety-labeled comments are markedly more negative than others, whereas every other dimension shows negligible effect sizes.
- **Polarization outweighs solidarity:** Polarizing comments constitute 32.4% of the discourse, compared to only 5.8% solidarity comments.
- Engagement favors solidarity: Despite being fewer, solidarity comments receive more likes, replies, and higher sentiment scores, indicating stronger positive audience engagement.
- Voices are concentrated: A smaller group of users drives more polarizing content.
- Echo chambers confirmed: The solidarity share shows a flat, stationary trend, suggesting a stable echo chamber – those who express solidarity keep doing so at consistent rates, bouncing back when their presence dips.
- **Polarizing behavior also stable:** The polarizing share is similarly flat and even less volatile, pointing to a core group of persistent polarizers.
- **No drift, no convergence:** Over time, the relative sizes of the solidarity and polarizing groups remain constant, neither converging nor diverging oscillating around fixed means.
- Key takeaway: Climate discussions show no sign of becoming more unifying or divisive. Instead, they
  reflect enduring filter bubbles where each camp maintains its position a persistent dual-bubble dynamic.



## Thank you!

